

Interpreting mm-wave sounder observations over deep convection

Ziad S. Haddad, Randy Sawaya, Ousmane Sy, Svetla Hristova-Veleva, F Joe Turk JPL/Caltech

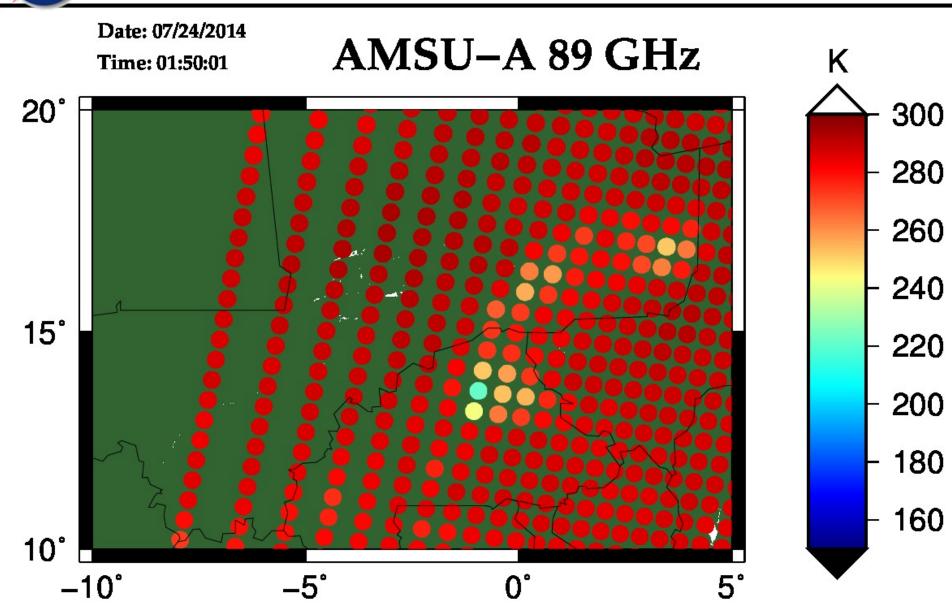
High-frequency sounders **can resolve convection**, so can we retrieve **convection intensity**

instead of surface rain?

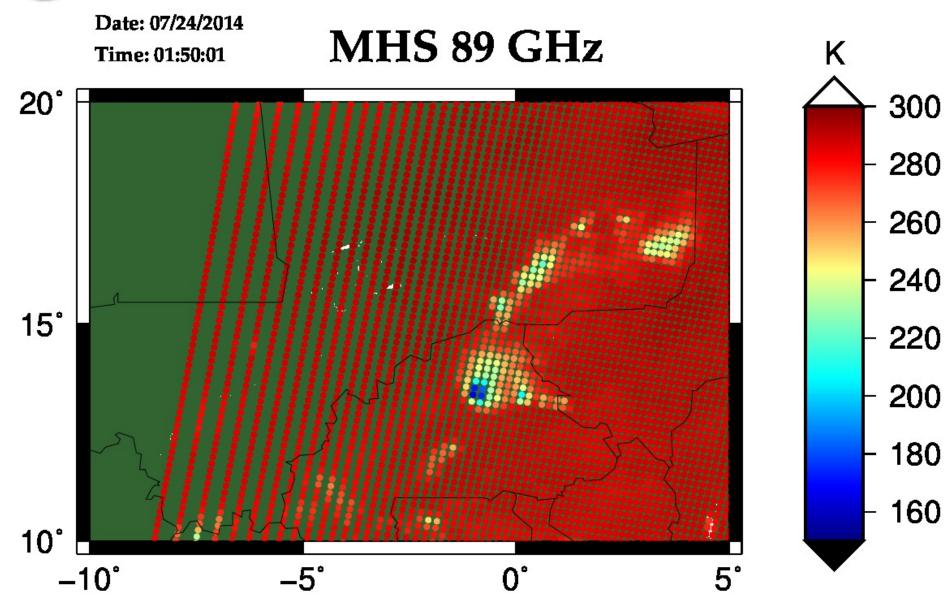
What are the biggest difficulties for the forward modeling? (hydrometeors)
What are the biggest difficulties for assimilation? (", biases, H₂O vapor)
Measurements don't amount to many independent pieces of info

Applications in addition to forecasting (diagnostics)

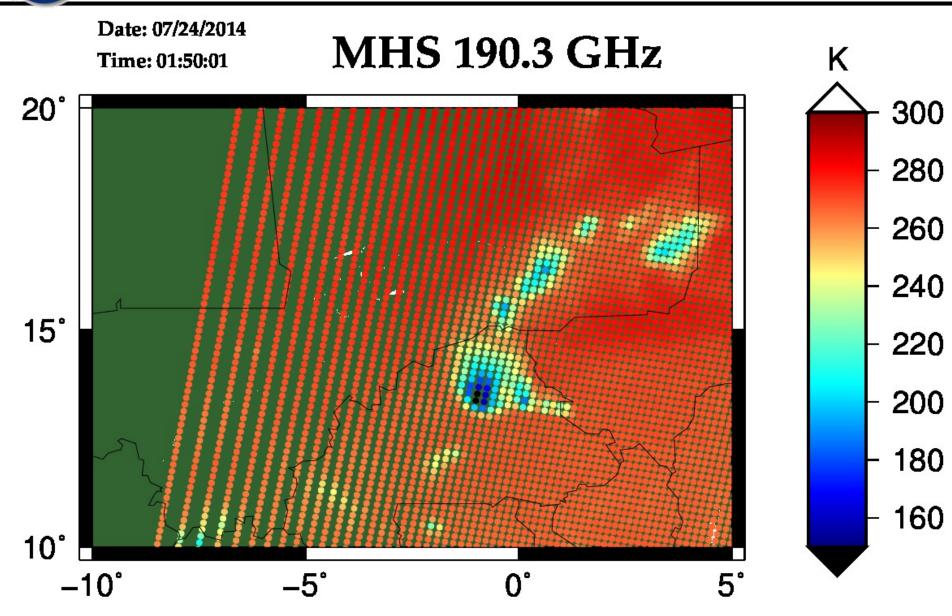




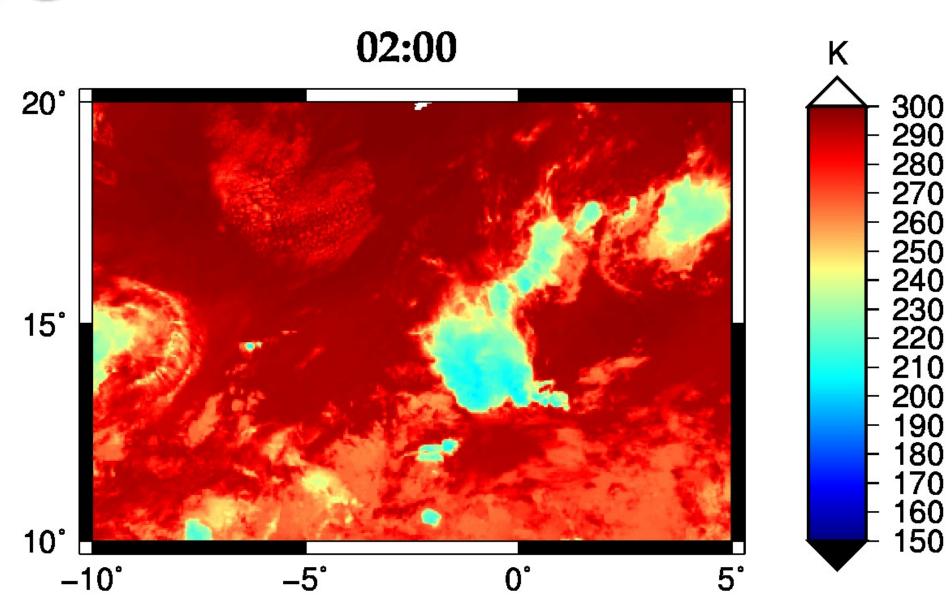














⇒ with radiometer (MHS):

- 1. Use a database of nearly-simultaneous TRMM radar + MHS obs
- 2. Trust the radar as the truth relative to Condensed Water
- 3. Use sample conditional mean to estimate vertical PC of CW from $T_{\rm b}$, and to estimate mean $T_{\rm b}$ associated to CW
- 4. Use CRM simulations to fit (T_b conditional mean) to the (model) water vapor

Oth step: detection

linear discriminant biased locally

 \Rightarrow add non-linear variables – for MHS, these are

$$T_6 = k_6 \frac{T_{186} - T_{184}}{T_{186} + T_{184}} \quad T_7 = k_7 \frac{T_{190} - T_{184}}{T_{190} + T_{184}} \quad T_8 = k_8 \frac{T_{190} - T_{186}}{T_{190} + T_{186}}$$

linear discriminant with non-linear variables works right out of the box



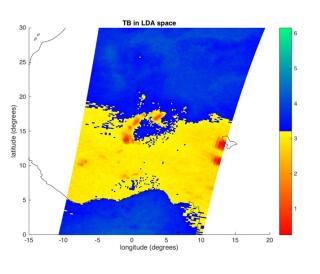
⇒ Semi-empirical approach with MHS:

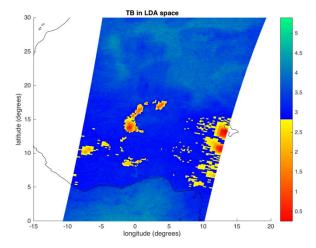
Oth step: detection (using database of coincident radar+radiom obs) linear discriminant biased locally

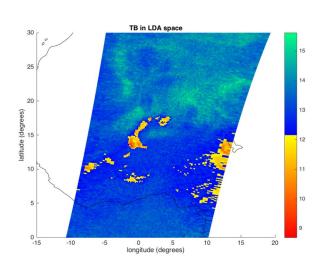
 \Rightarrow add non-linear variables – for MHS, these are

$$T_6 = k_6 \frac{T_{186} - T_{184}}{T_{186} + T_{184}} \quad T_7 = k_7 \frac{T_{190} - T_{184}}{T_{190} + T_{184}} \quad T_8 = k_8 \frac{T_{190} - T_{186}}{T_{190} + T_{186}}$$

linear discriminant with non-linear variables works right out of the box









⇒ Semi-empirical approach with MHS:

1. Use a database of nearly-simultaneous radar + radiometer obs

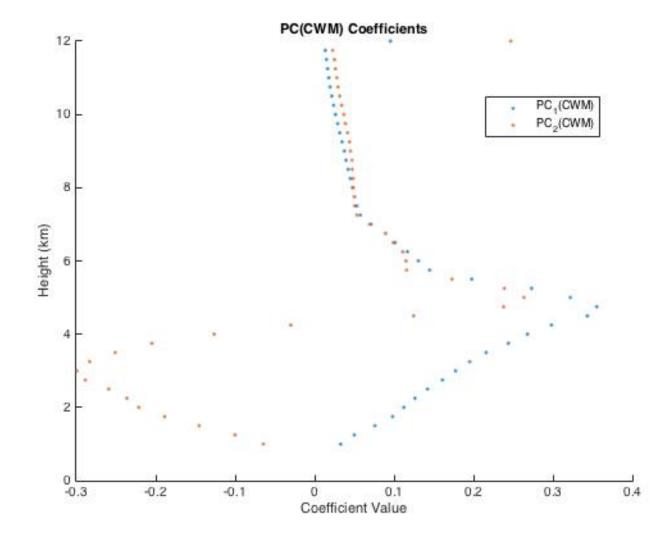
Oth step: detection using linear discriminant with non-linear variables

 $0'^{th}$ step: In the rain, use clear-air cPC₂, cPC₃, cPC₄, ... instead of T_b themselves

Then use in-rain $rPC_1(cPC_2, cPC_3, cPC_4, ...)$ & $rPC_2(cPC_2, cPC_3, cPC_4, ...)$ instead of T_b themselves

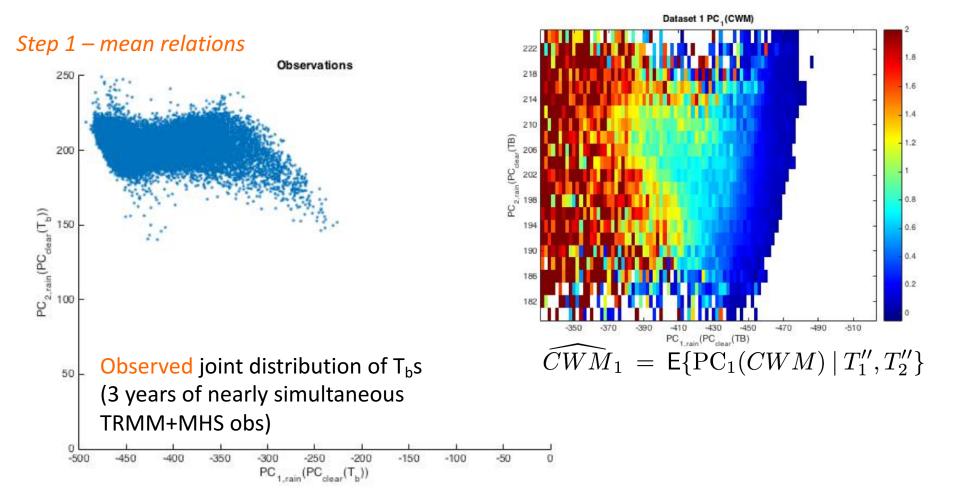
(and similarly use vPC1(CW) and vPC2(CW) instead of vertical profile of CW)



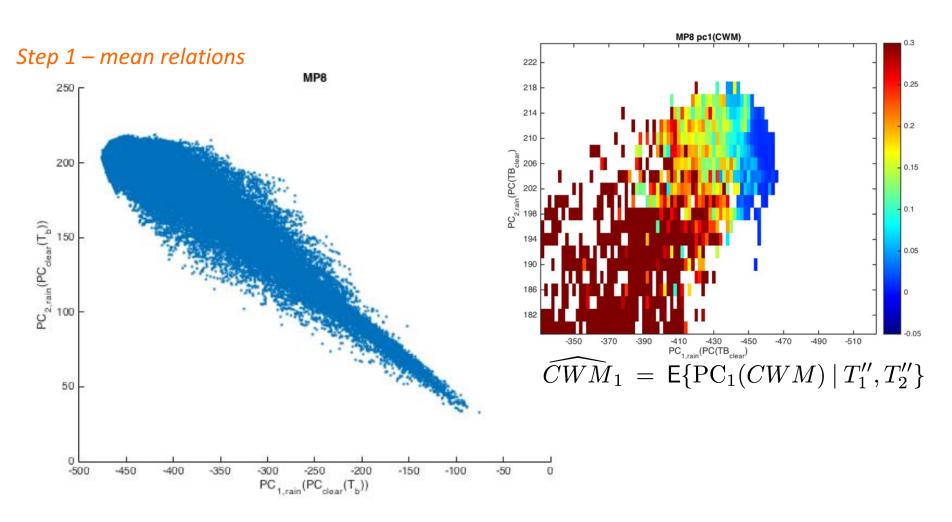


vPC1(CW) and vPC2(CW)



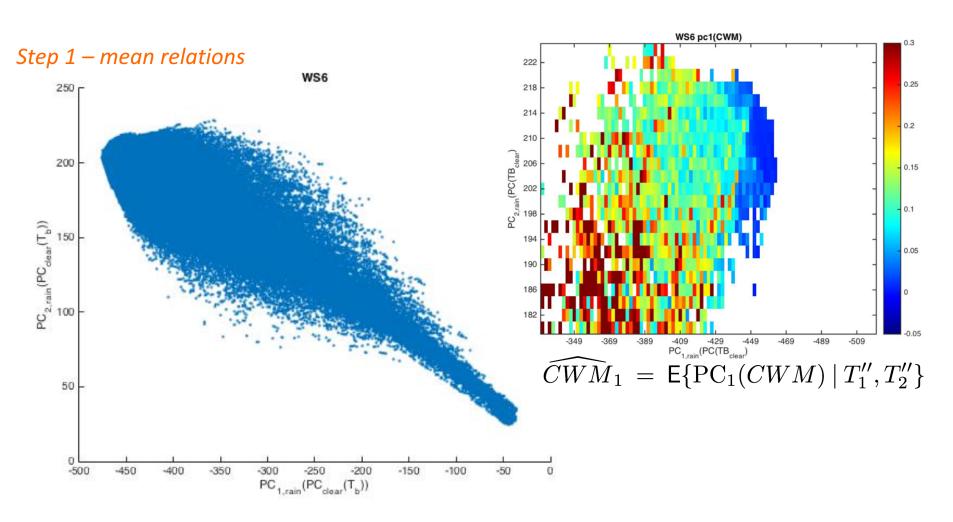






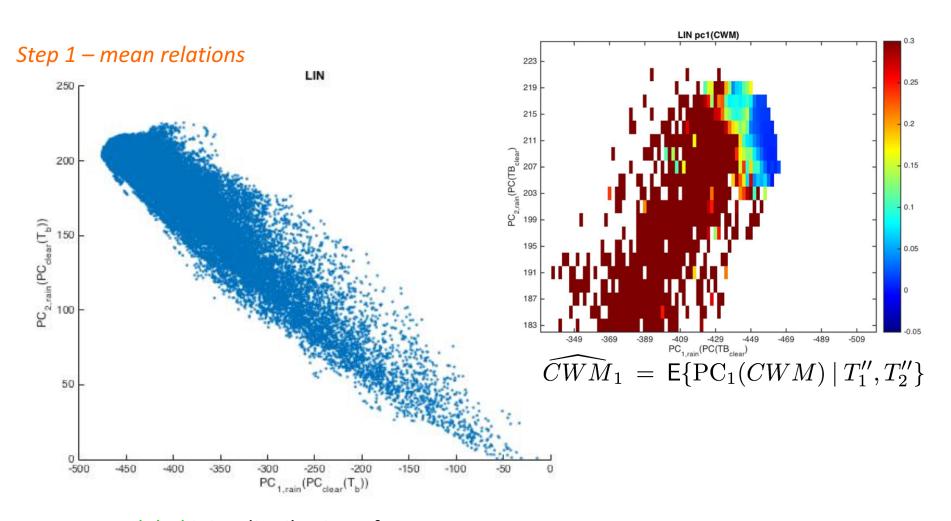
Modeled joint distribution of T_bs 10 consecutive instants (10mins apart) of Isabel





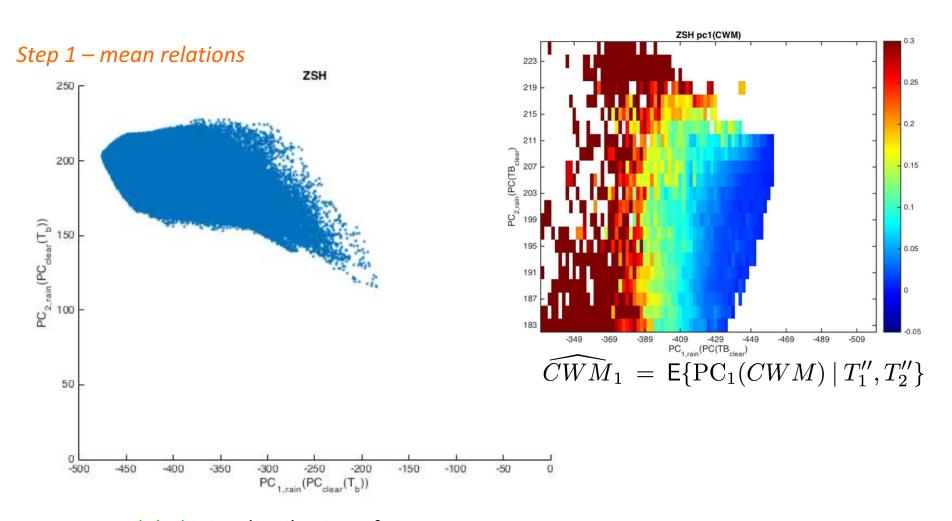
Modeled joint distribution of T_bs 10 consecutive instants (10mins apart) of Isabel





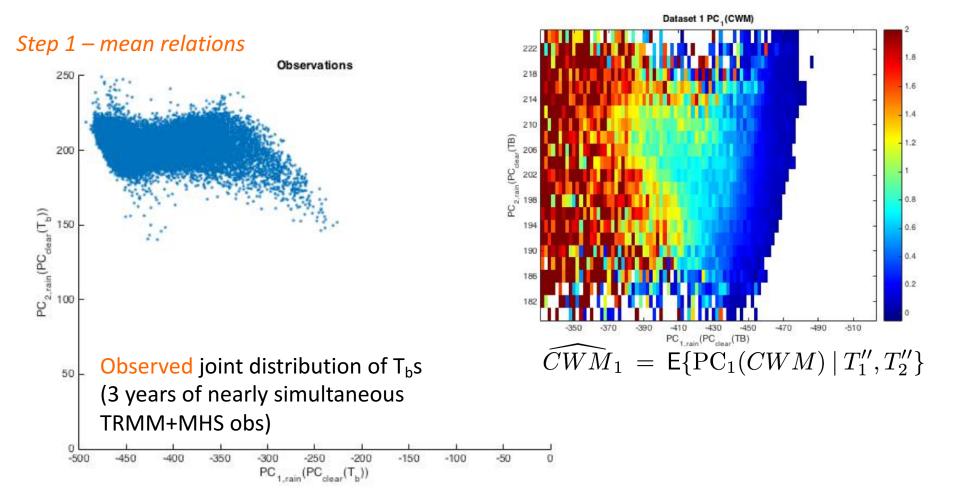
Modeled joint distribution of T_bs 10 consecutive instants (10mins apart) of Isabel





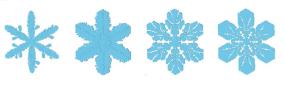
Modeled joint distribution of T_bs 10 consecutive instants (10mins apart) of Isabel

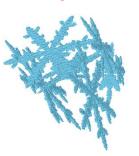






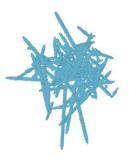
What is "OOSy-zsh" Hydrometeor representation?











Schmitt & Heymsfield 2010: for individual hydrometeor,

mass = a
$$D_{max}^{b}$$

with 0.004 < a < 0.009 and 1.8 < b < 2.4

So: use 32 sample (a,b) and sort Kuo's synthetic hydrometeors into one of the 32 classes

Then sample (μ, Λ) , and for each sample value in each (a,b) class, calculate D_{MaWe} and σ_{MaWe}

Finally, keep only those for which $0.2 < D_{MaWe}/CW^{0.17} < 2 \quad \text{and} \quad 0.15 < \sigma_{MaWe}/D_{MaWe}^{1.3} < 0.6$

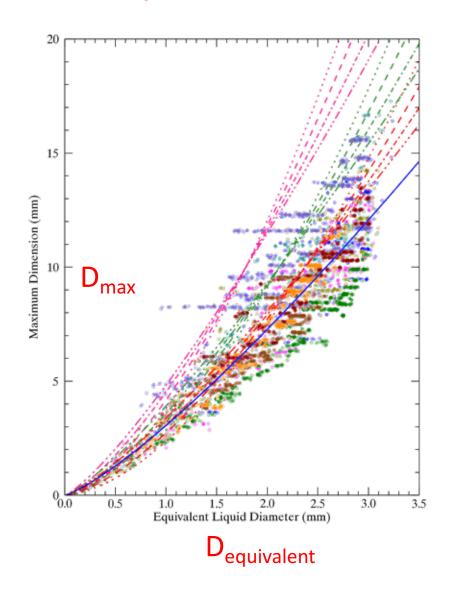


What is "OOSy-zsh" Hydrometeor representation?

Schmitt & Heymsfield 2010:

mass = $a D_{max}^{b}$ with 0.004 < a < 0.009 and 1.8 < b < 2.4

expect: $D_{max} = [\pi/(6a)]^{1/b} (Dequiv)^{3/b}$





⇒ non-empirical part of Semi-empirical approach:

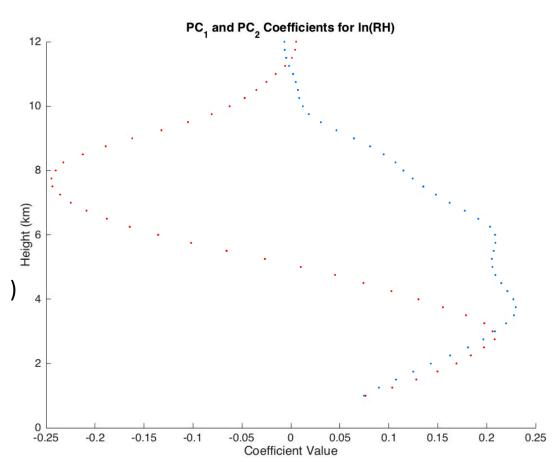
Step 2 – departure from mean

Accepting the empirical mean relation between T and CW,

use model sims to express dependence of residual

$$\Delta T'' = (T'')_{\text{MHS}}$$
$$-E(T'' | PC_1(CW), PC_2(CW)$$

on water vapor

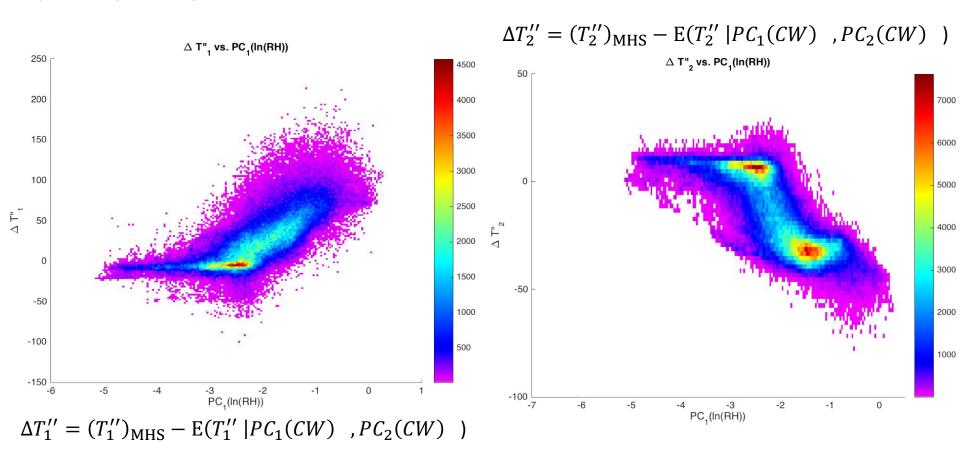


 \Rightarrow look for correlation between ΔT and vPCs of RH ...



\Rightarrow look for correlation between ΔT and vPCs of RH:

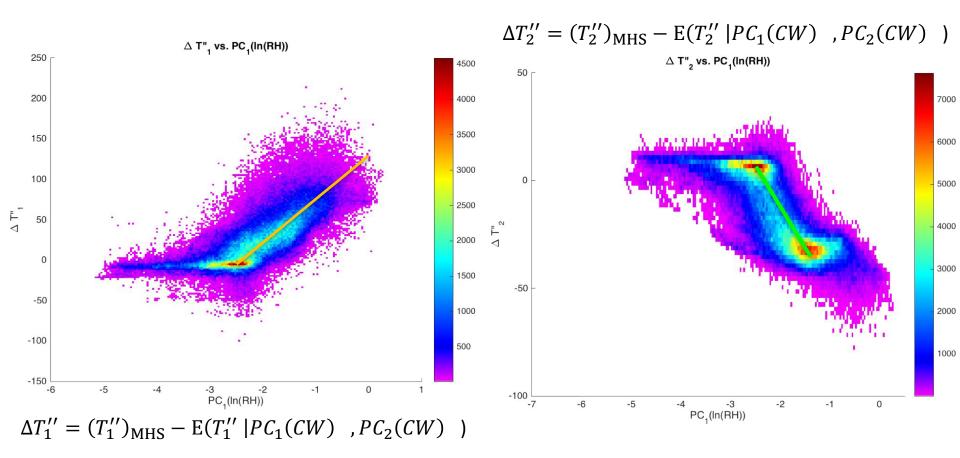
Step 2 – departure from mean





\Rightarrow look for correlation between ΔT and vPCs of RH:

Step 2 – departure from mean



Simulation shows that $T - (\cdot \mid CW)$ is indeed correlated with RH!

empirically consistent mm-wave signatures



⇒ Semi-empirical approach (both steps combined):

$$T_1'' = \mathrm{E}(T_1'' | PC_1(CW), PC_2(CW)) + \mathrm{E}(\Delta T_1'' | PC_1(RH))$$

$$T_2'' = \mathrm{E}(T_2'' | PC_1(CW), PC_2(CW)) + \mathrm{E}(\Delta T_2'' | PC_1(RH))$$
from database of coincidences from model simulations with

